	Data Cl	eaning													
[10]:	<pre>df.drop(['Status','unnamed1'],axis=1,</pre>				inplace=True)								Λ Ψ	古 무	Î
[10]:		User_ID	Cust_name	Product_ID	Gender	Age Group	Age	Marital_Status	State	Zone	Occupation	Product_Category	Orders	Amount	
	0	1002903	Sanskriti	P00125942	F	26-35	28	0	Maharashtra	Western	Healthcare	Auto	1	23952.0	
	1	1000732	Kartik	P00110942	F	26-35	35	1	Andhra Pradesh	Southern	Govt	Auto	3	23934.0	
	2	1001990	Bindu	P00118542	F	26-35	35	1	Uttar Pradesh	Central	Automobile	Auto	3	23924.0	
	3	1001425	Sudevi	P00237842	М	0-17	16	0	Karnataka	Southern	Construction	Auto	2	23912.0	
	4	1000588	Joni	P00057942	М	26-35	28	1	Gujarat	Western	Food Processing	Auto	2	23877.0	
	11246	1000695	Manning	P00296942	М	18-25	19	1	Maharashtra	Western	Chemical	Office	4	370.0	
	11247	1004089	Reichenbach	P00171342	М	26-35	33	0	Haryana	Northern	Healthcare	Veterinary	3	367.0	
	11248	1001209	Oshin	P00201342	F	36-45	40	0	Madhya Pradesh	Central	Textile	Office	4	213.0	
	11249	1004023	Noonan	P00059442	М	36-45	37	0	Karnataka	Southern	Agriculture	Office	3	206.0	
	11250	1002744	Brumley	P00281742	F	18-25	19	0	Maharashtra	Western	Healthcare	Office	3	188.0	
1	11251 rc	ows × 13 c	olumns												

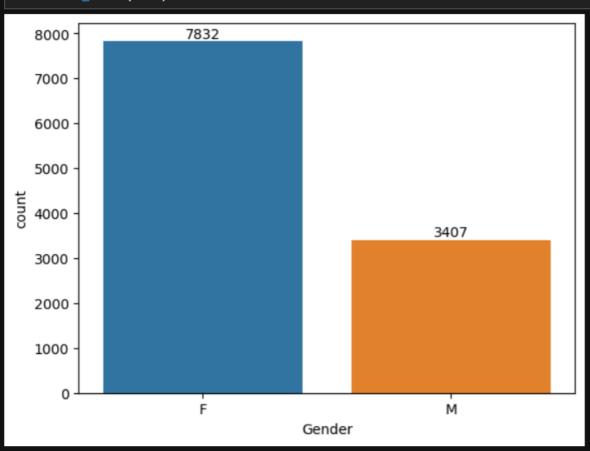
```
[16]: pd.isnull(df).sum()
[16]: User_ID
                           0
      Cust_name
                           0
      Product_ID
                           0
      Gender
                           0
      Age Group
                           0
      Age
                           0
      Marital_Status
                           0
      State
                           0
      Zone
                           0
      Occupation
                           0
      Product_Category
                           0
      Orders
                           0
      Amount
                          12
      dtype: int64
      df.dropna(inplace=True)
[19]:
      pd.isnull(df).sum()
[20]:
[20]: User_ID
                          0
      Cust_name
                          0
      Product ID
                          0
      Gender
                          0
      Age Group
                          0
      Age
                          0
      Marital_Status
                          0
      State
                          0
      Zone
                          0
      Occupation
                          0
      Product_Category
                          0
      Orders
                          0
      Amount
                          0
      dtype: int64
```

Exploratory Data Analysis

plotting a barchart for Gender using seaborn

```
[49]: ax = sns.countplot(x= 'Gender', hue = 'Gender', data=df )

for bars in ax.containers:
    ax.bar_label(bars)
```

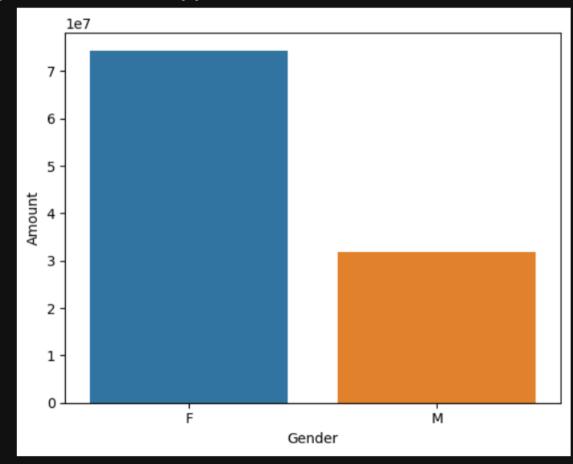


plotting a bar chart Gender vs Total

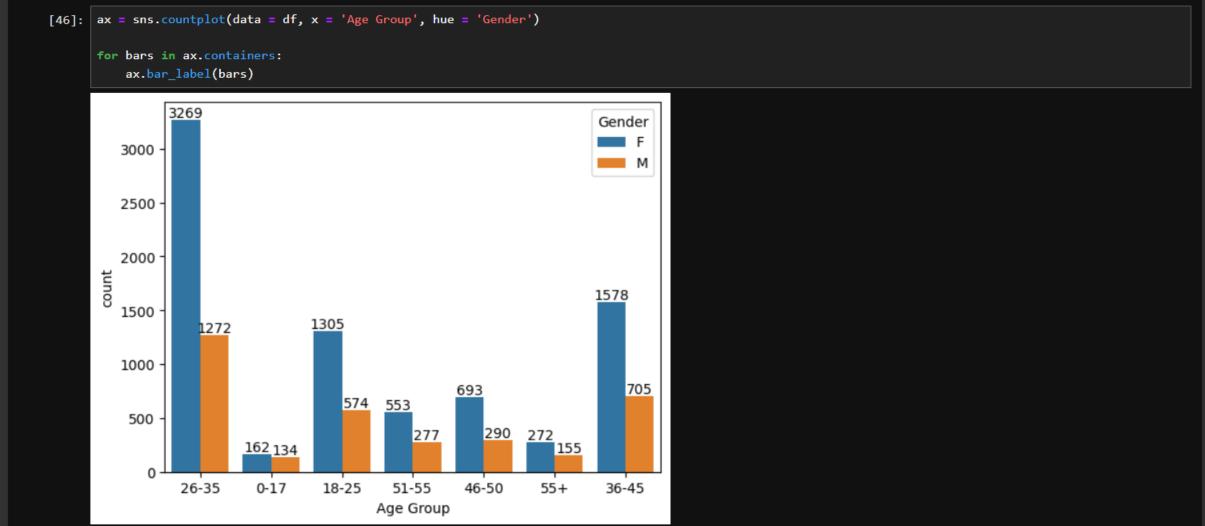
```
[47]: sales_gen = df.groupby(['Gender'],as_index=False)['Amount'].sum().sort_values(by='Amount', ascending=False)

sns.barplot(x= 'Gender', y='Amount', hue = 'Gender', data=sales_gen)
```

[47]: <Axes: xlabel='Gender', ylabel='Amount'>



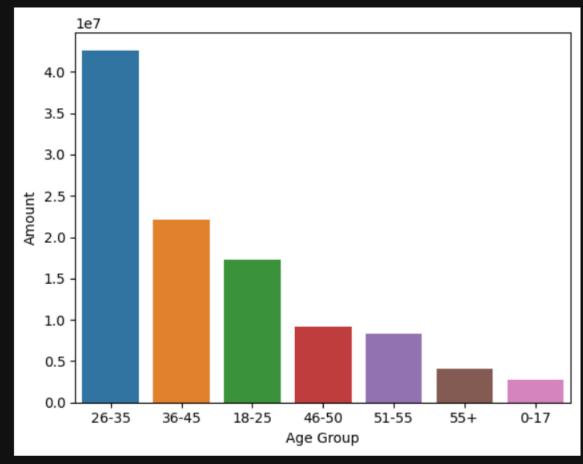
From above graphs we can see that most of the buyers are female and even the purchasing power of females are greater than men



```
[52]: sales_age = df.groupby(['Age Group'],as_index=False)['Amount'].sum().sort_values(by='Amount', ascending=False)

sns.barplot(x= 'Age Group', y='Amount', data=sales_age, hue ='Age Group')
```

[52]: <Axes: xlabel='Age Group', ylabel='Amount'>

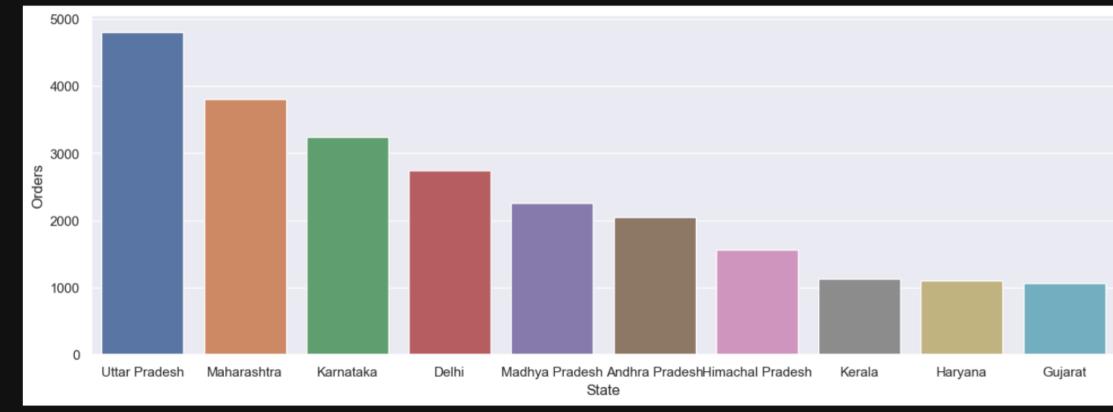


From above graph we can see that most of the buyers are of age group between 26-35 years Females

```
[59]: sales_state = df.groupby(['State'], as_index = False)['Orders'].sum().sort_values(by='Orders', ascending=False).head(10)

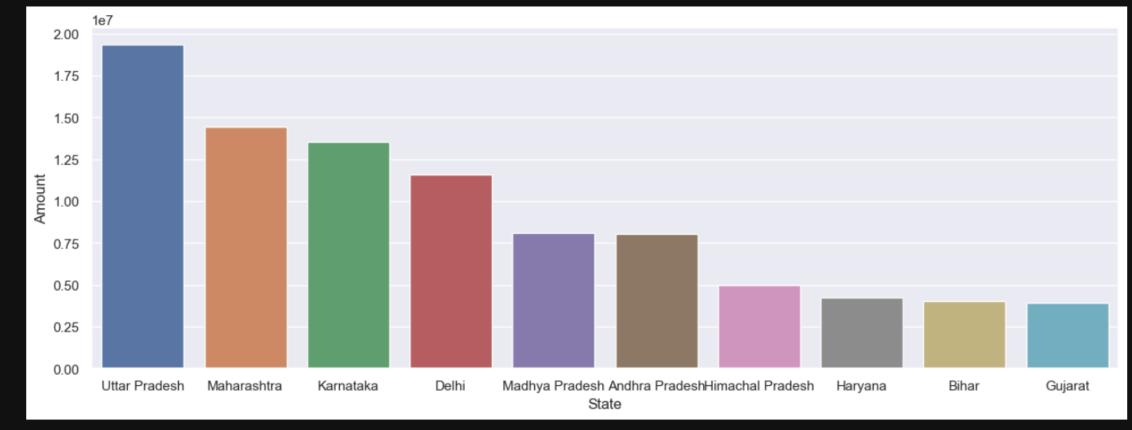
sns.set(rc={'figure.figsize':(15,5)})
sns.barplot(data = sales_state, x='State', y='Orders', hue ='State')
```



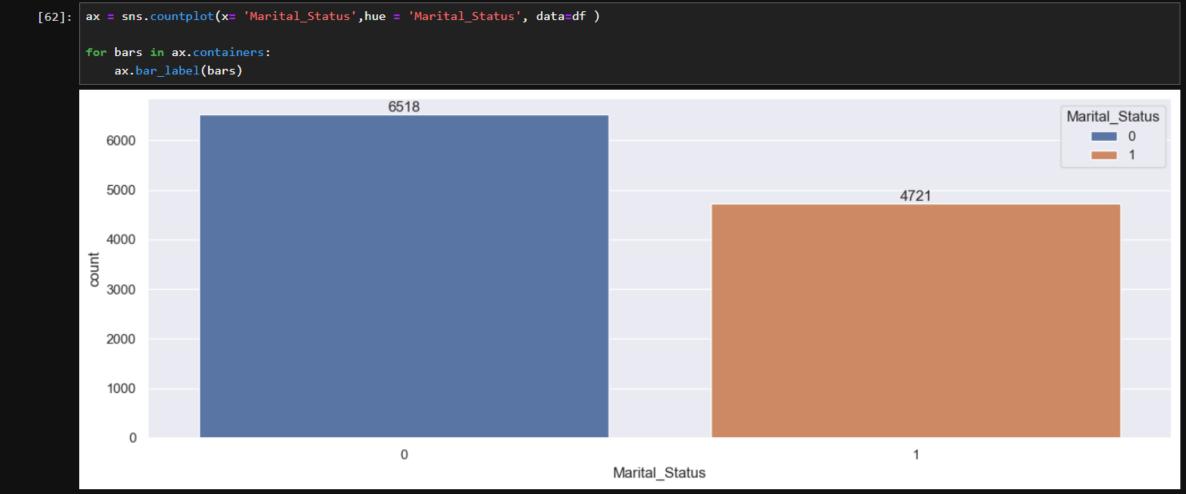


```
[60]: sales_state = df.groupby(['State'], as_index = False)['Amount'].sum().sort_values(by='Amount', ascending=False).head(10)
sns.set(rc={'figure.figsize':(15,5)})
sns.barplot(data = sales_state, x='State', y='Amount', hue ='State')
```

[60]: <Axes: xlabel='State', ylabel='Amount'>



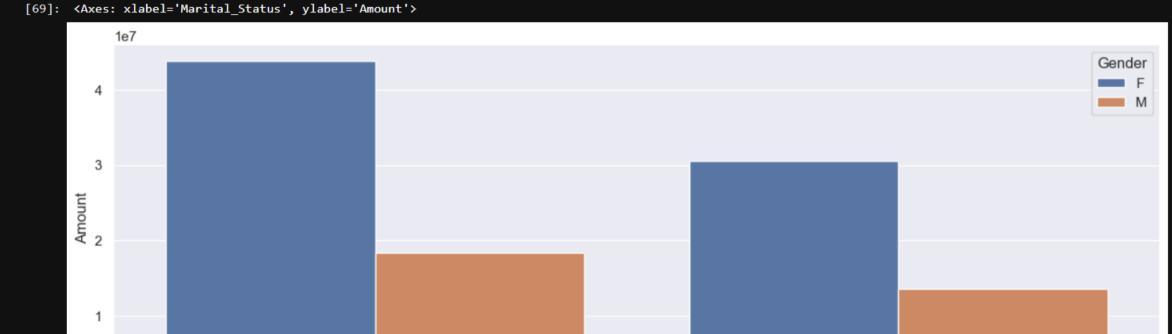
From above graphs we can see that most of the orders and total sales amount are from Uttar Pradesh, Maharashtra and Karnataka respectively



```
sales_state = df.groupby(['Marital_Status','Gender'], as_index = False)['Amount'].sum().sort_values(by='Amount', ascending=False)

sns.set(rc={'figure.figsize':(15,5)})

sns.barplot(data = sales_state, x='Marital_Status', y='Amount', hue ='Gender')
```



Marital\_Status

From above graph we can see that most of the buyers are married (women) and they have high purchasing power

0



Media

Lawyer

Banking

Occupation

Retail

IT Sector

Aviation

Hospitality

Textile

Chemical

Agriculture

From above graph we can see that most of the buyers are working in IT, Healthcare and Aviation sector

Food Processing

200

Healthcare

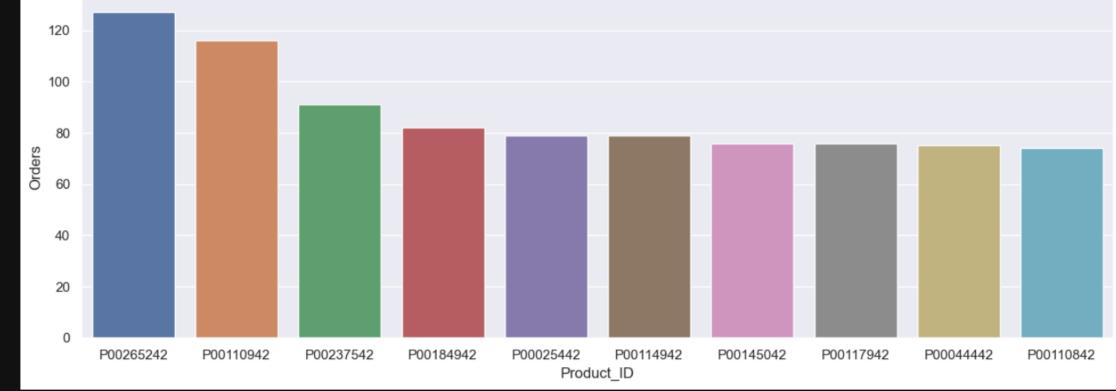
Govt

Automobile



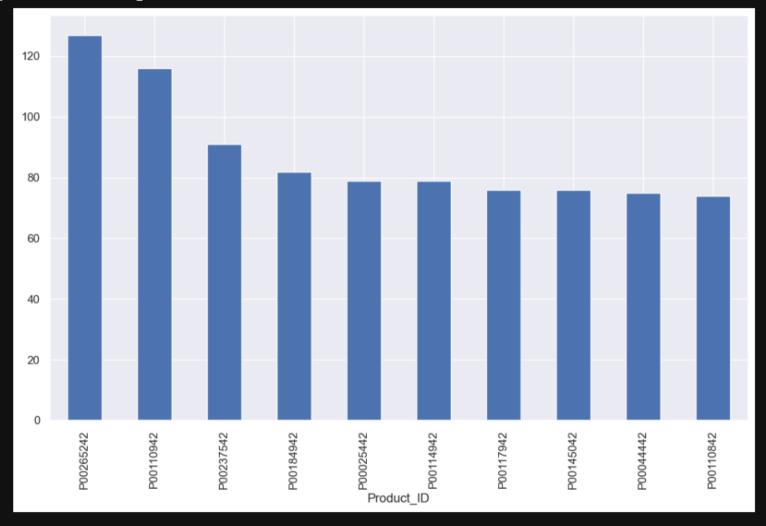
```
[85]: sales_state = df.groupby(['Product_ID'], as_index = False)['Orders'].sum().sort_values(by='Orders', ascending=False).head(10)
sns.set(rc={'figure.figsize':(15,5)})
sns.barplot(data = sales_state, x='Product_ID', y='Orders', hue = 'Product_ID')

[85]: <Axes: xlabel='Product_ID', ylabel='Orders'>
```



[86]: fig1, ax1 = plt.subplots(figsize=(12,7))
df.groupby('Product\_ID')['Orders'].sum().nlargest(10).sort\_values(ascending=False).plot(kind='bar')

[86]: <Axes: xlabel='Product\_ID'>



## Conclusion:

Married women age group 26-35 years from UP, Maharashtra, Karnataka working in IT, Healthcare and Aviation sector are more likely to buy products from food, clothing and electronic category